

# VAA-Weekly-Progress

09/05 - 09/10

# Context

- Last week, we reviewed some papers on semi-supervised keypoint localization and the visual transformer pose estimator
- We decided for this week to focus on solidifying a baseline model to proceed with, and comparing models pretrained on different datasets

# Deciding on a Baseline System

- OpenMMLab?
- HRNet vs. RTMPose?
  - Read?
  - Performance?
    - Established benchmark datasets (e.g., AP-10k, AnimalPose)
    - Try your own images
      - What may cause trouble for pose estimators?
- Pre-trained on AP-10k or pre-trained AnimalPose?
  - What are the benefits and drawbacks?
  - Why not both?
  - What does the dataset evaluation system look like?
    - How did AP-10k tackle the detection problem?

# Deciding on a Baseline System

- What do you choose?
- Why do you choose it?
- What is needed to get it working?
- What could go wrong?

# HRNET Evaluation

Trained on Animal Pose (w32 config):

Pros:

- More accurate
  - Placement of wither kp on body focused pictures (on a higher place)
- Marks both neck and wither
- More accurate kp placement for average size blurry images

Cons:

- Excessive number of kp's on partially occluded objects
- Excessive/lack of kp's on weird poses
- Misplaces head kp's in blurry night time pictures

Trained on AP-10k (w48 config)

Pros:

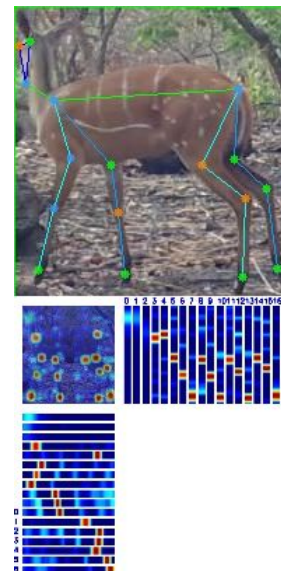
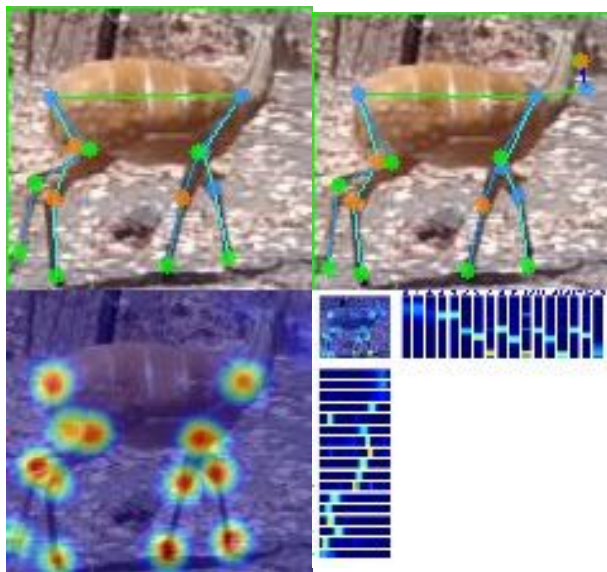
- More accurate kp's for
  - Night time images (black and white)
  - Weird poses
  - Tracks hip kp more accurately in partial images
  - Small blurry images

Cons:

- Does not see the head in blurry night time images
- Adds additional hip kp in night time images

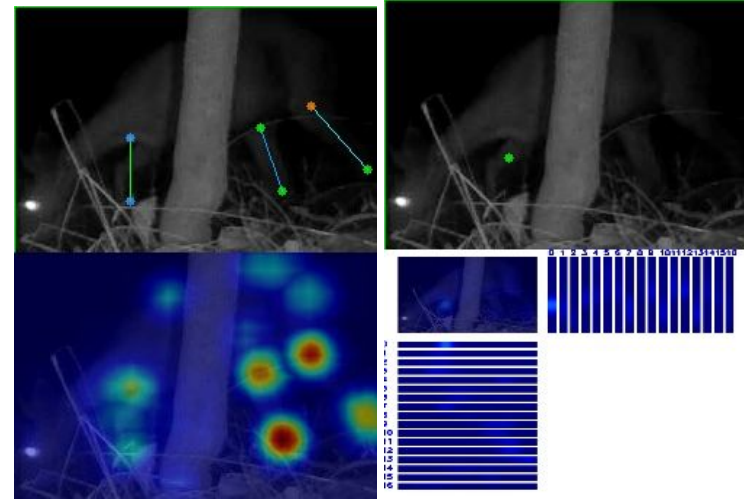
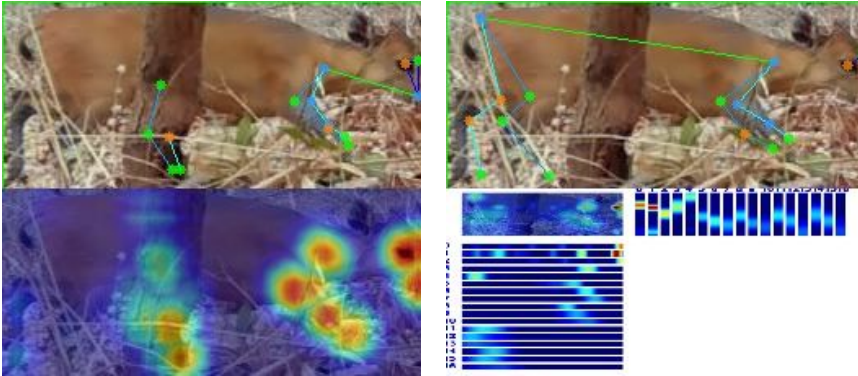
# HRNet(w48) vs RTMPose

We ran both HRNet and RTMPose (pretrained on AP-10k) on our sample of antelope data. It appears that when predicting keypoints on the same image, RTMPose does a better job on detecting keypoints on the head.



# HRNet(w48) vs RTMPose

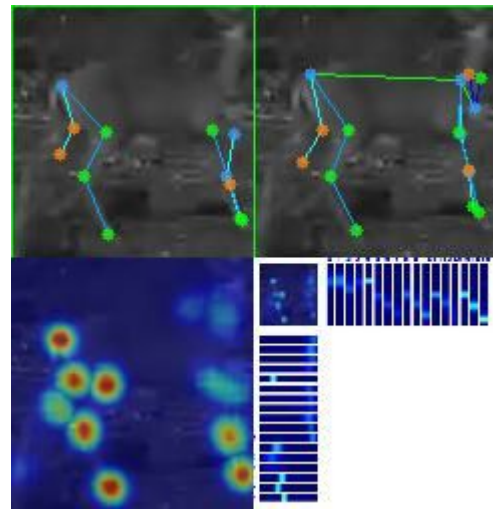
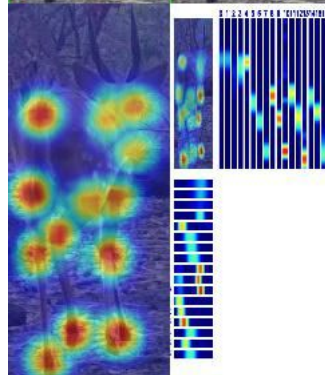
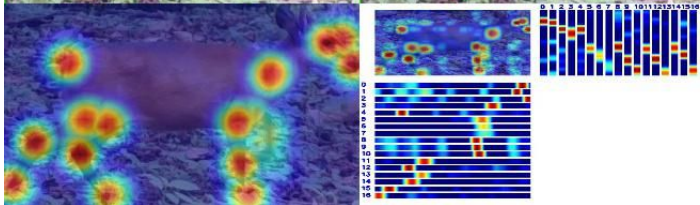
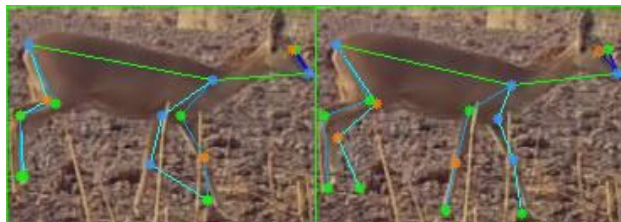
When detecting keypoints on an image with some obstruction, both models have difficulty. Note: in other night-time images, the two models perform at similar levels.





# HRNet(w48) vs RTMPose

RTMPose seems to be slightly more robust, but both models suffer from the same challenges





# RTMPose Evaluation

Trained on AP-10k:

Pros:

- Tends to work well for when detecting the head using kps
- Cut off images are labeled fairly well
- Twisted/Weird poses still have relatively accurate kps
- Some night time images have ideal kps

Cons:

- Struggles with obstacles blocking animal of interest
- Some blurry night images lack kps in necessary areas

# Final Thoughts

We decided to proceed with RTMPose (trained on AP-10K) as our baseline model, because based on a qualitative analysis with a small subset of our data, it appears to have better performance compared to HRNet, specifically in images with occlusion and overall does a better job of identifying keypoints.

# Next Steps

- Train RTMPose on Animal Pose data set and compare and contrast qualitative performance to determine on which training set results in better qualitative performance of RTMPose
- Combine Animal Pose and AP-10K into one point (converting wither and neck points in Animal Pose to neck in AP-10K) and train RTMPose on the combined data set and test qualitative performance against RTMPose trained on each data set separately
- Consider possible use of animal data sets as training data sets (AP-36K)

# Personal Progress

# Medha

- Annotated 100 images with labels (Set 4)
- Familiarized with OpenMMPose and analyzed outputs of two different models (HRNet and RTMPose) on our antelope data
- Working on setup for OpenMMPose on personal deadcat
- Met with team to decide on the baseline model to proceed with

# Parth

- Annotated 100 images with labels (Set 6)
- Ran and analyzed RTMPose and HRNet Models trained AP-10K dataset
- Tested the models on our data from Week 1
- Decided on baseline model with team

# Aryan

- Annotated images from the set 0 with labels
- Ran both models on previous data for a comparative analysis



# Josh

- Annotated 100 images with labels (Set 1)
- Read abstracts and introductions of HRNet and RTMPose papers and watched talks to become familiar with the two models architectures and general advantages and disadvantages
- Analyzed HRNet and RTMPose on set 1 of antelope images to qualitatively determine which network is more suited to be our baseline system
- Discussed with team about which model to choose as baseline

# Claire

- Annotated 100 images with labels (Set 2)
- Familiarized with OpenMMPose, and analyzed and compared results of running sorted antelope images with HRNet and RTMPose
- Discussed with team about which model to choose as baseline

# Armaan

- Annotated 100 images with labels (Set 5)
- Wrote script batch.py to run a specified model and configuration over a whole folder of antelope images at once.
- Analyzed HRNet trained on animal pose vs trained on AP-10k across different configurations (w32, w48, resnet101) on set 5 of antelope images to qualitatively determine which model (and configuration) is most suited to be our baseline system
- Discussed with team about which model to choose as baseline

# Shaan

- Labeled 100 images
- Re-used some of Armaan's code and put together a script to align outputs from pretrained HRNet and RTMPose models on our own antelope data
  - This streamlined our goal of conducting qualitative analysis as it allowed for viewing outputs on the same image side by side.